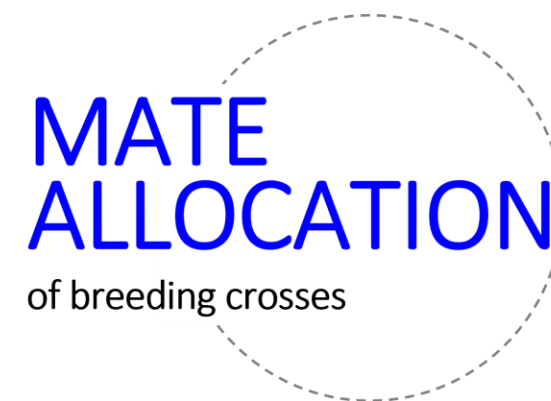




University of Florida
Sweet Corn and Potato Breeding and Genomics Lab
CANVAS – Salt Lake City, 2025



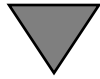
Enhancing Breeding Efficiency: Prediction and Optimization of Breeding Crosses for Long-term Genetic Gain and Diversity



Marco Antonio Peixoto* and Marcio Resende

- **Genomic selection**: the use of molecular markers and historical phenotypes to predict new unseen individuals.
- It has **revolutionized** animal/plant breeding programs.

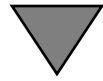
1. Selection accuracy
2. Shorten breeding cycles



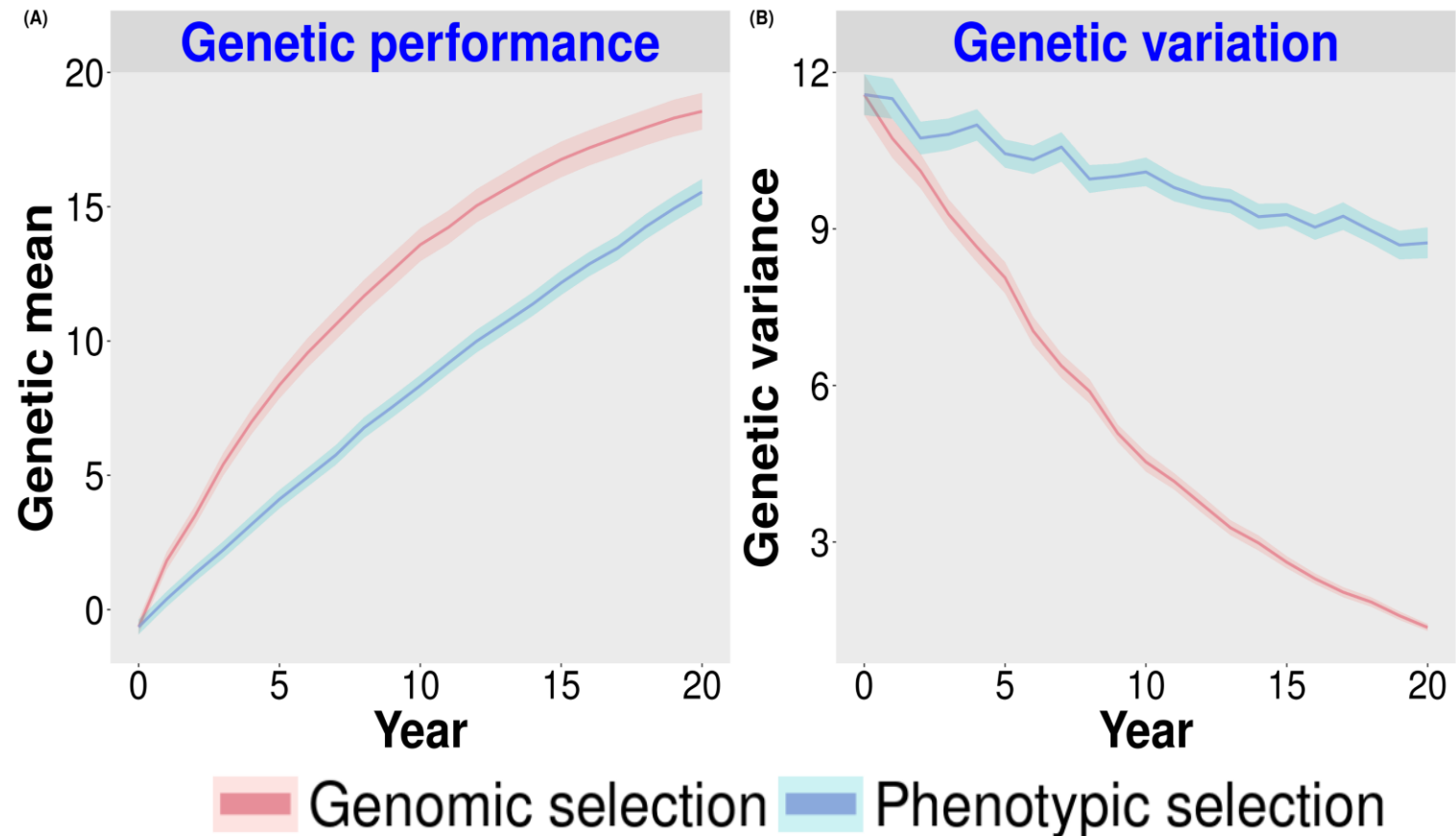
High performance

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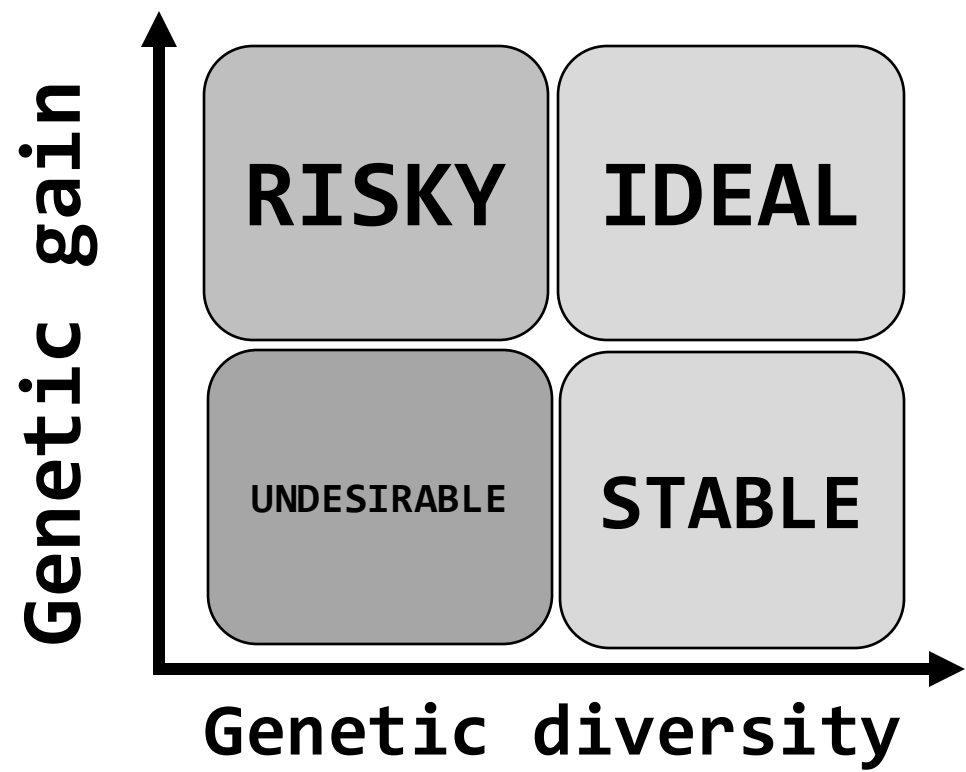
1. Selection accuracy
2. Shorten breeding cycles



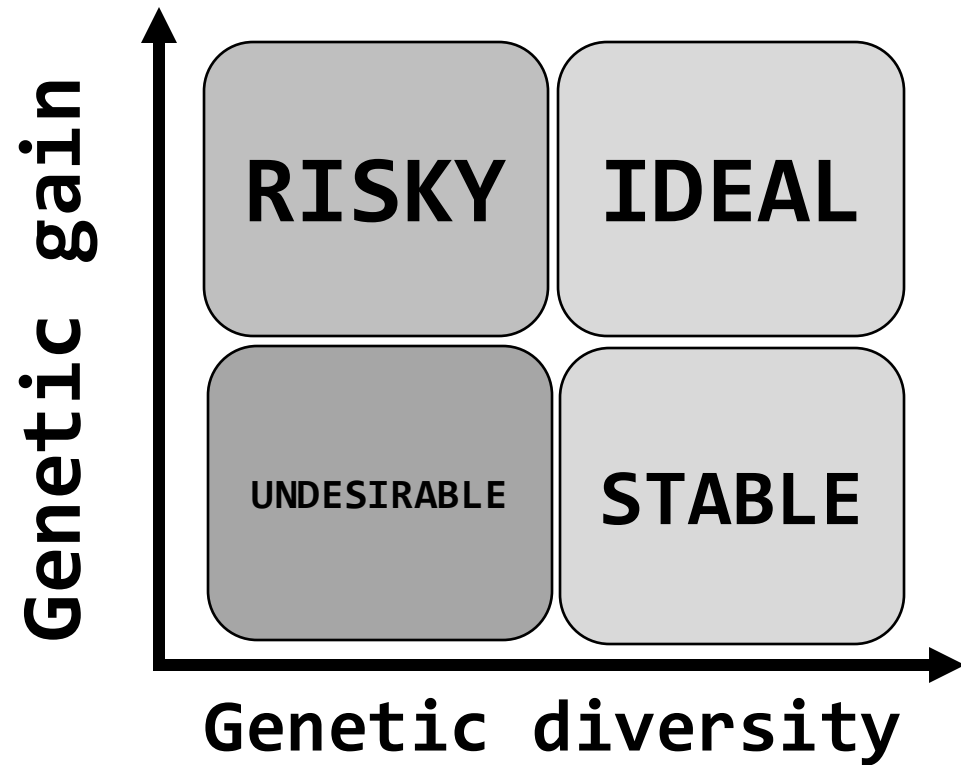
High performance



Problem

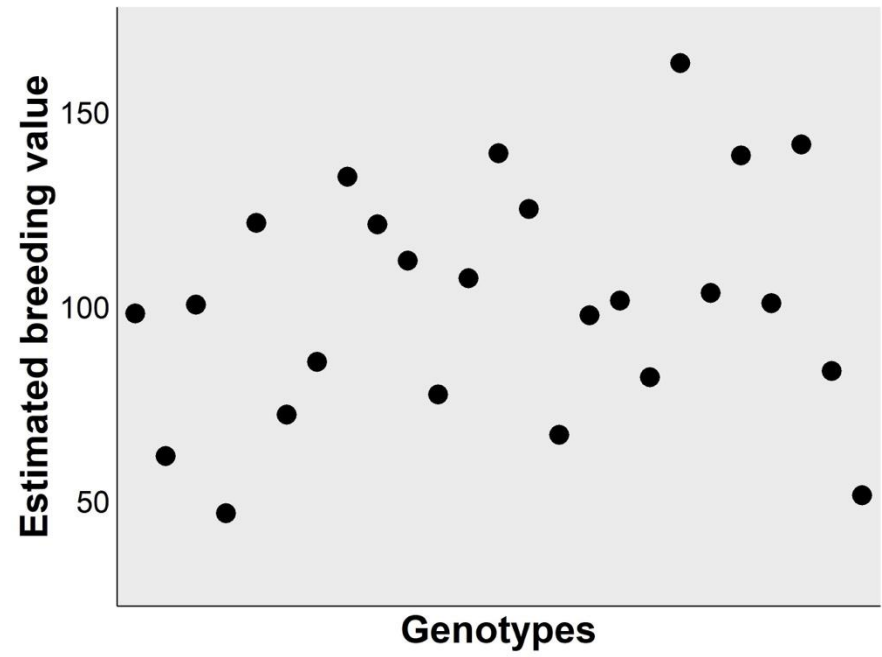


Problem

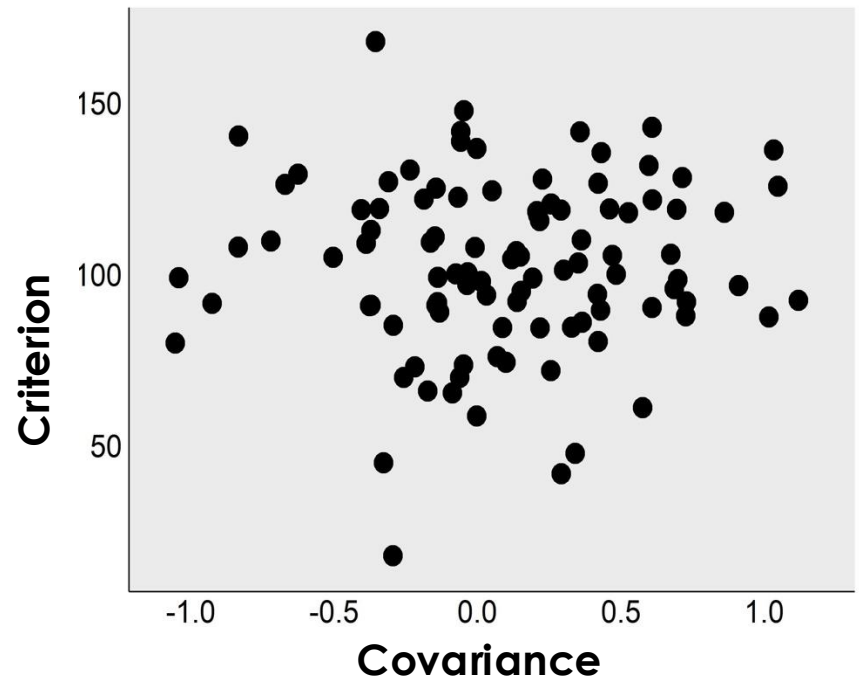
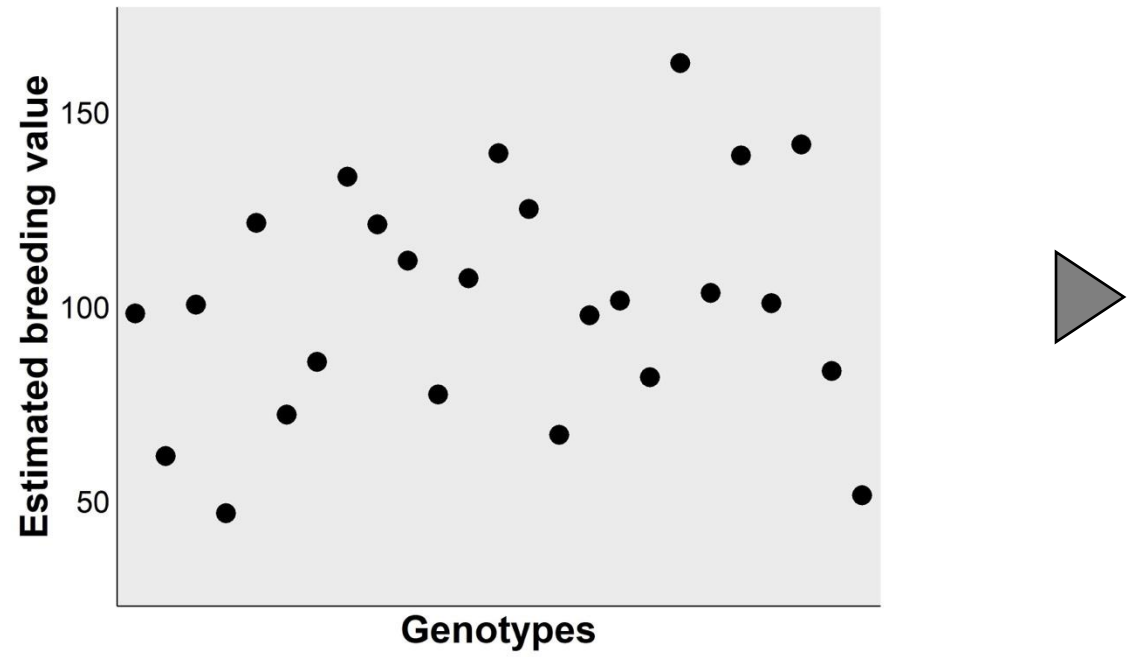


How do we balance rapid improvement with long-term sustainability in breeding programs?

Selection of parental combinations: optimal cross selection

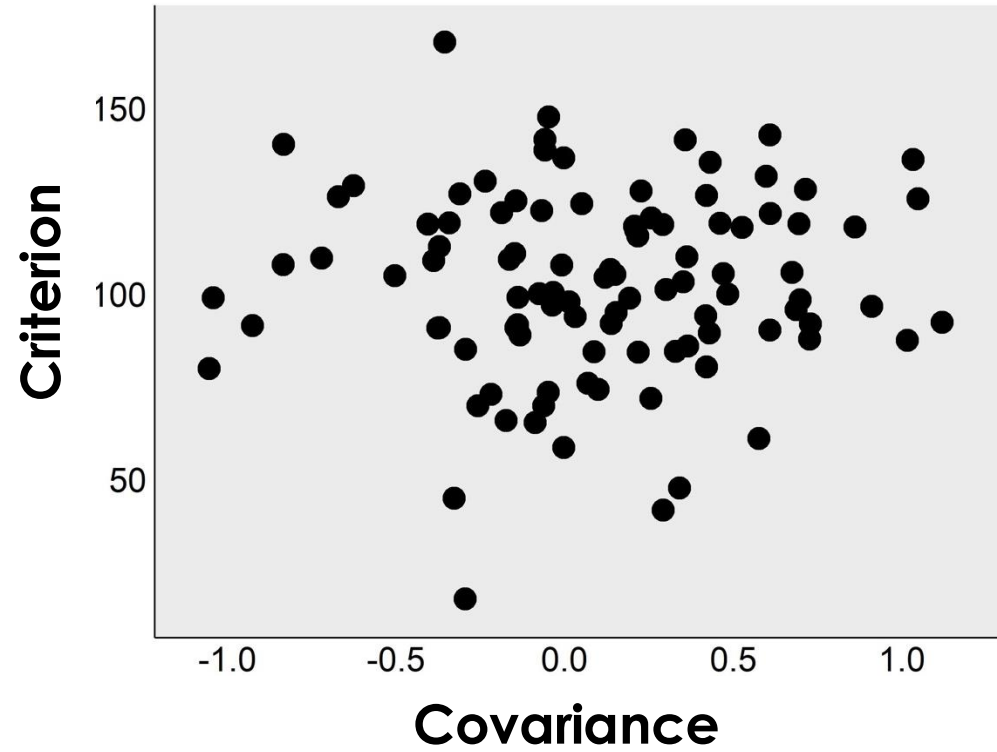


Selection of parental combinations: optimal cross selection

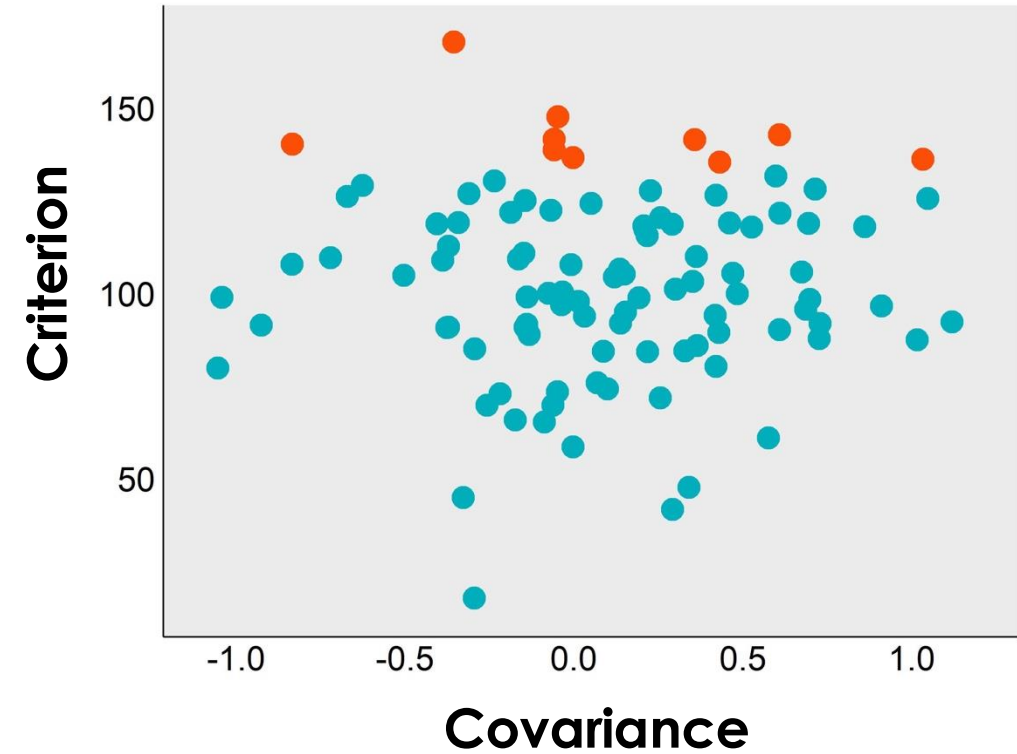
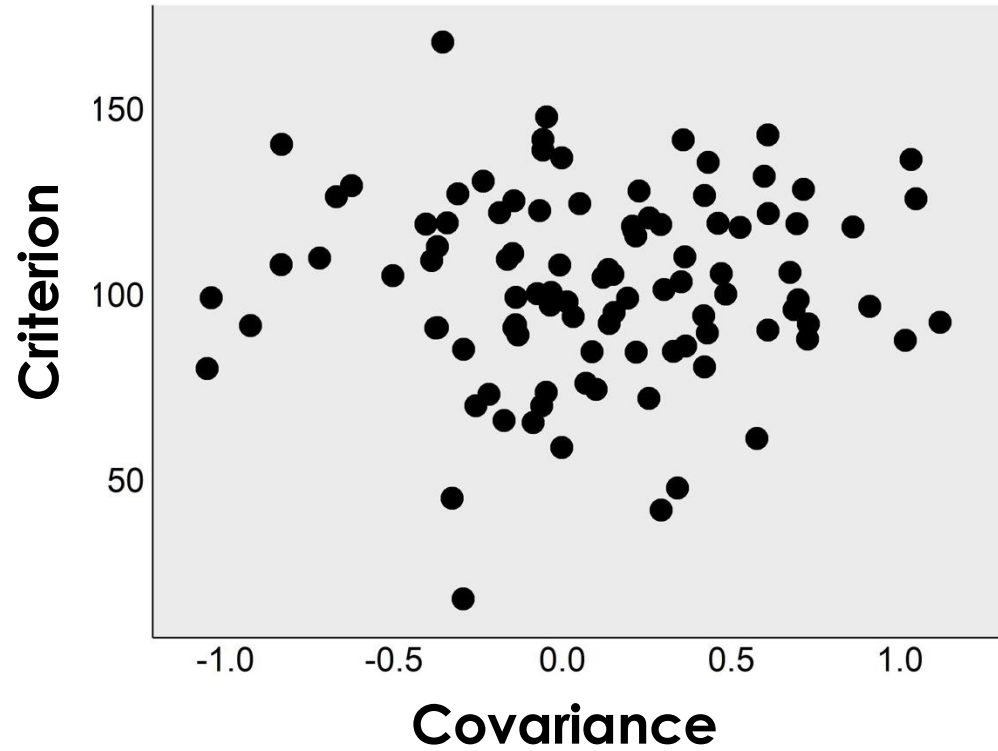


$$\mu_{ebv} = \frac{EBV_{p1} + EBV_{p2}}{2}$$

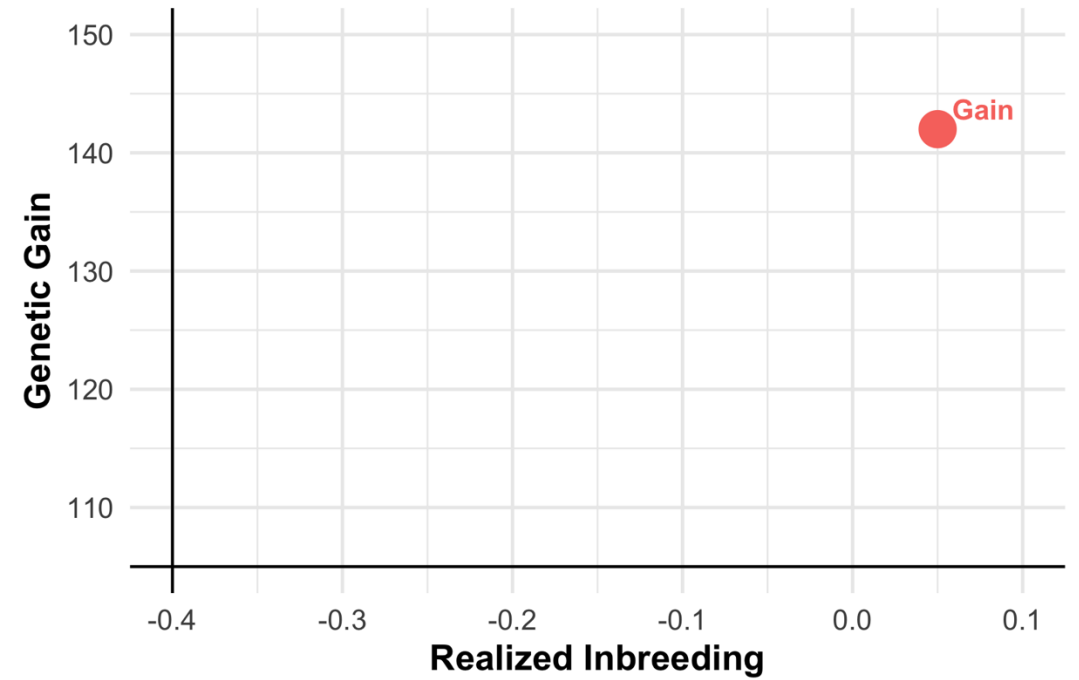
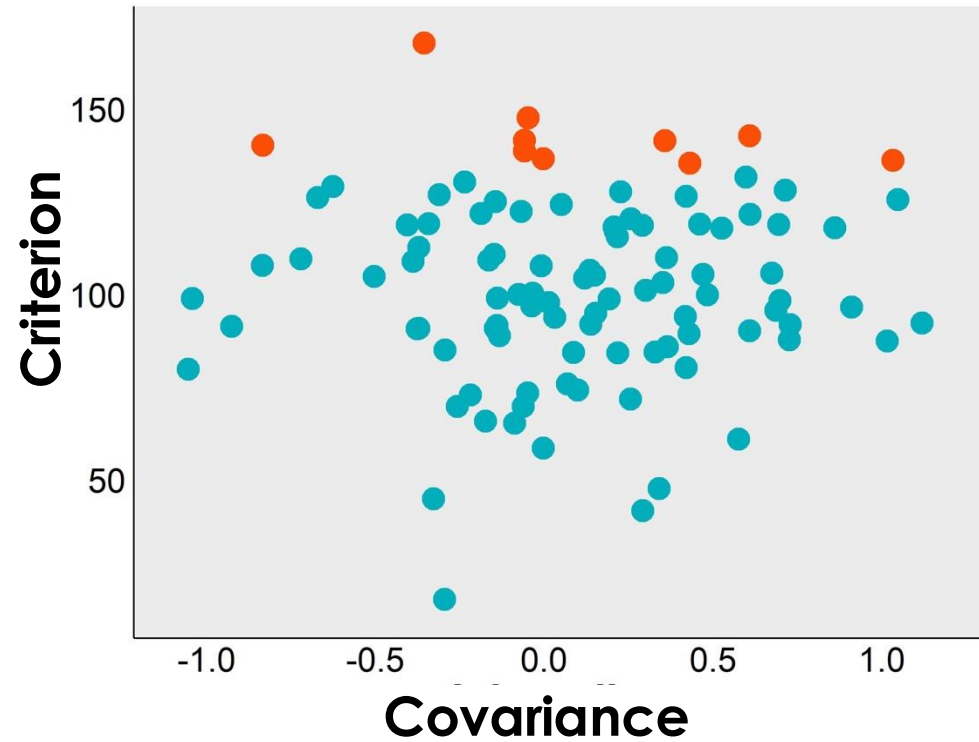
Optimal cross selection



Optimal cross selection

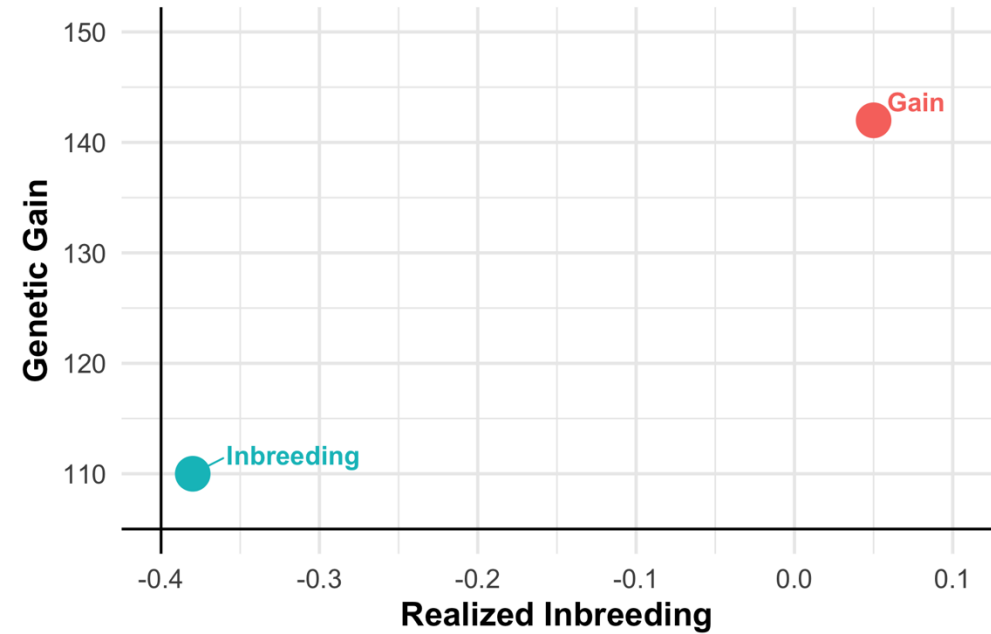
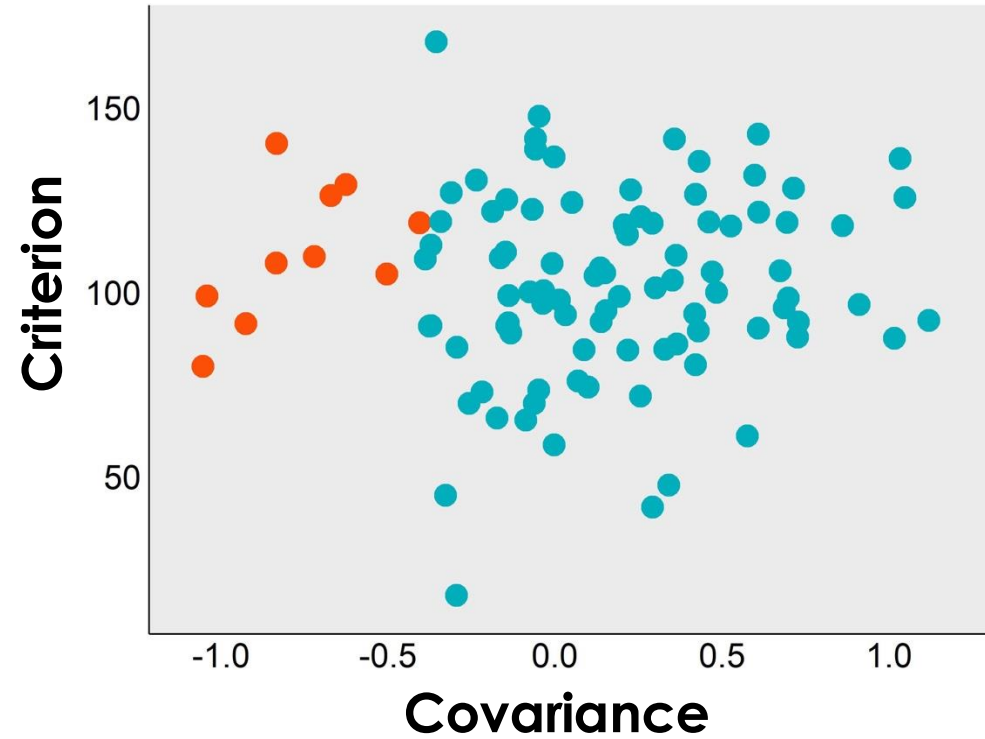


Optimal cross selection

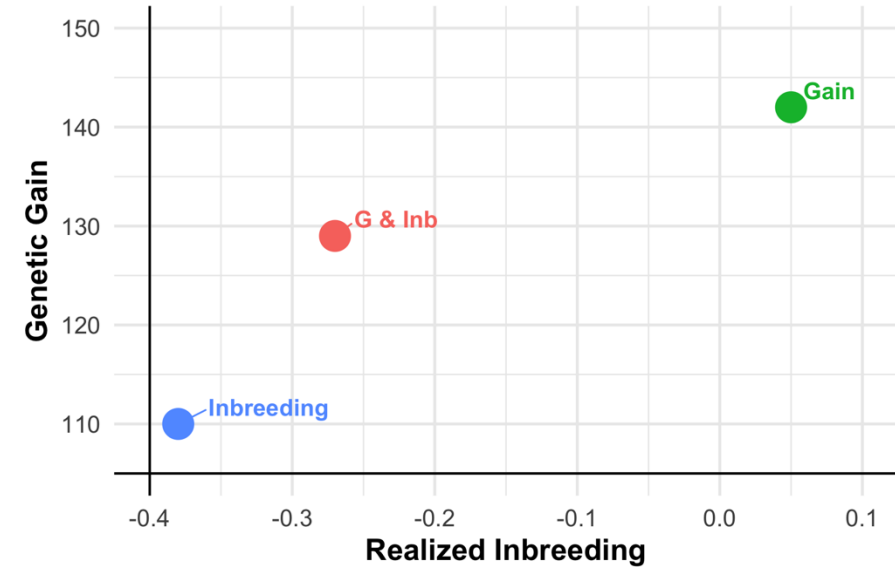
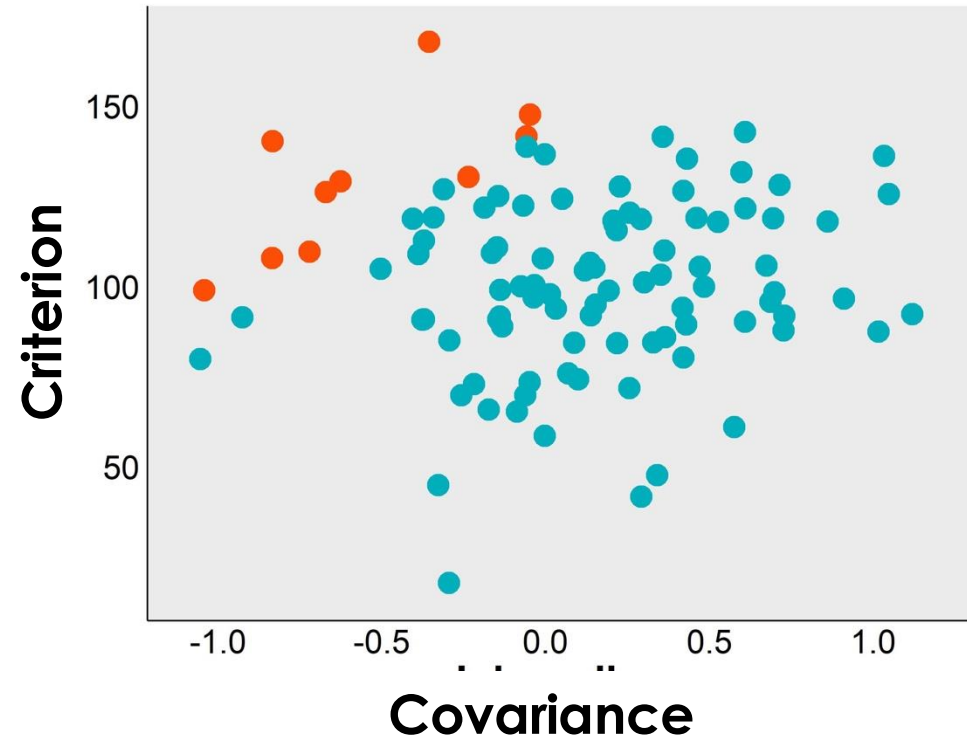


$$\frac{A}{2} = \text{Coancestry}$$
$$\text{Coancestry (t)} = \text{Inbreeding (t+1)}$$

Optimal cross selection



Optimal cross selection

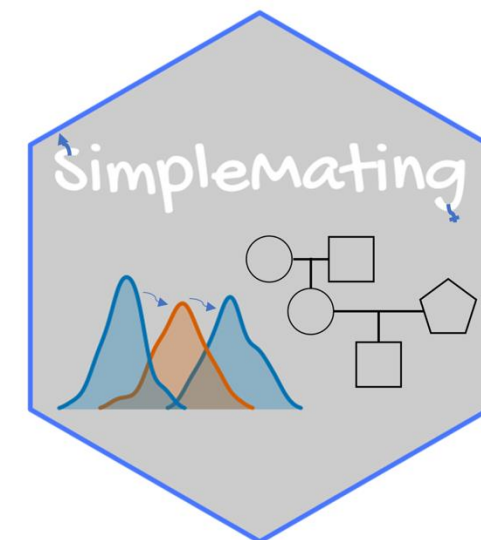


- Cross prediction and optimization

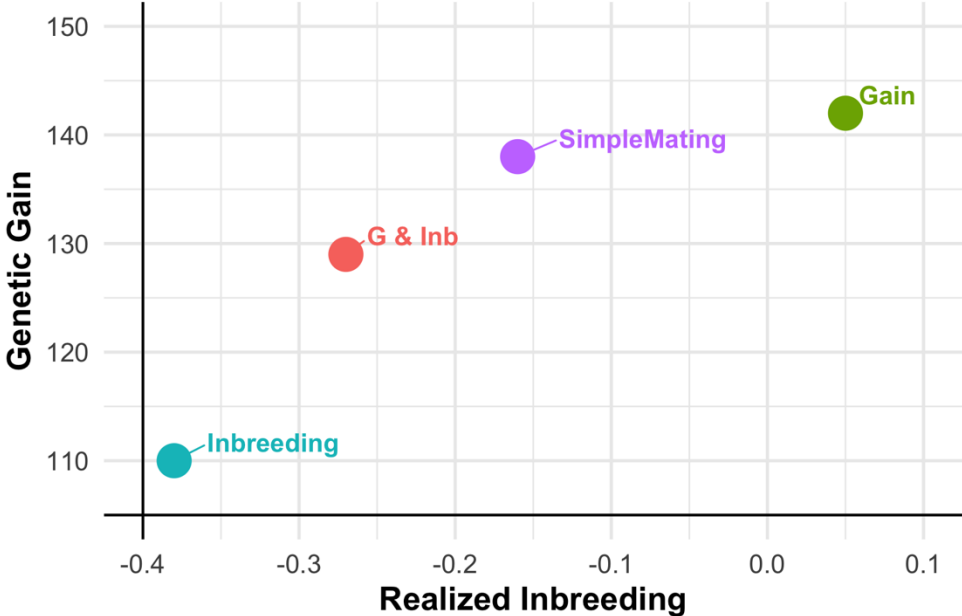
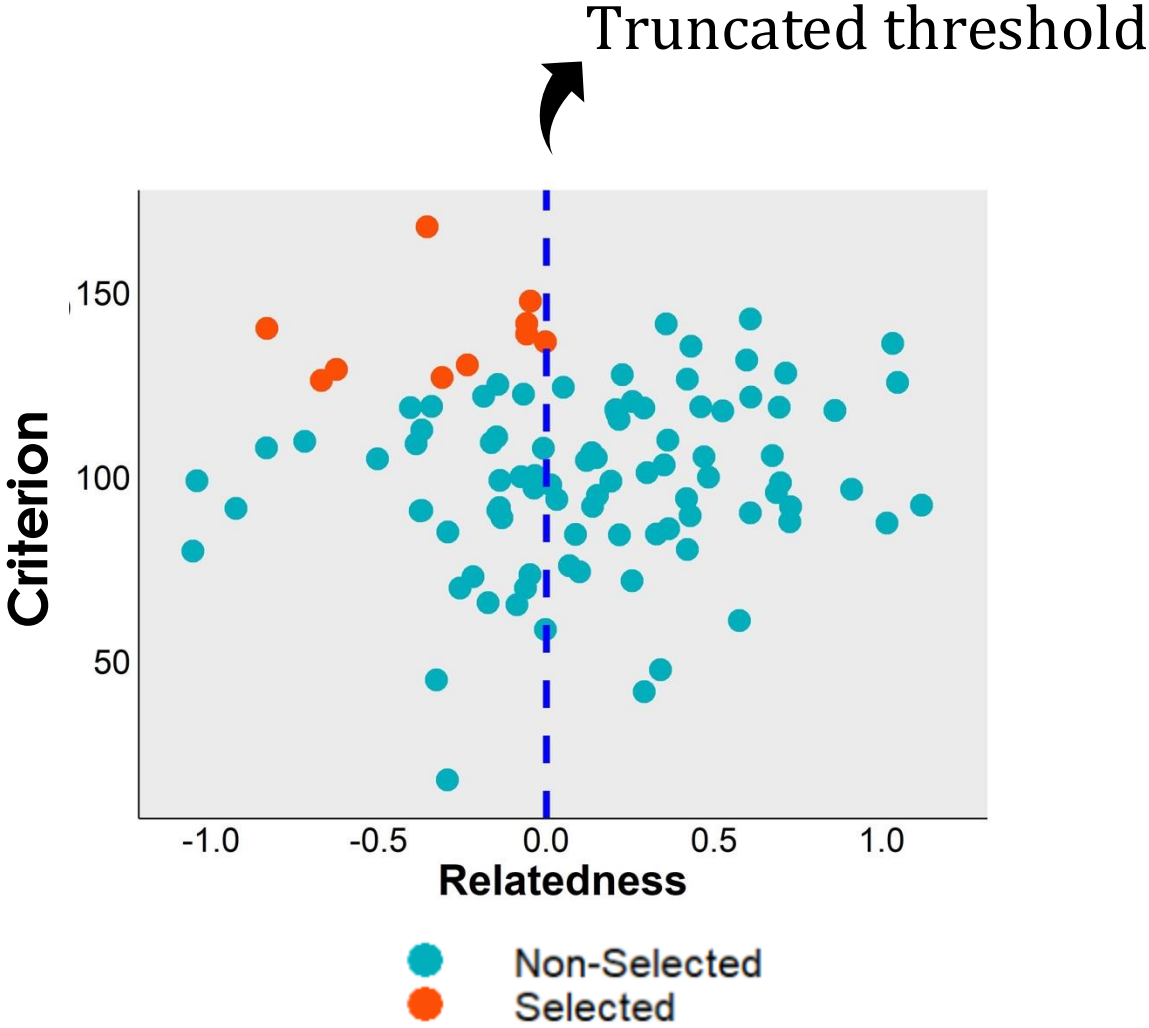
SimpleMating: R-package for prediction and optimization of breeding crosses using genomic selection

Marco Antônio Peixoto^{1,2}  | Rodrigo Rampazo Amadeu^{3,4}  | Leonardo Lopes Bhering¹  |
Luís Felipe V. Ferrão³  | Patrício R. Munoz³  | Márcio F. R. Resende Jr.² 

- Two module package
 - i. Prediction of cross performance
 - ii. Optimization of crosses



Optimal cross selection



SimpleMating: cross performance

1

Mid-parental value

$$\mu_{ebv} = \frac{ebv_{p1} + ebv_{p2}}{2}$$

3

Usefulness (A/AD)

$$UC = \mu + i\sigma$$

2

Total genetic value

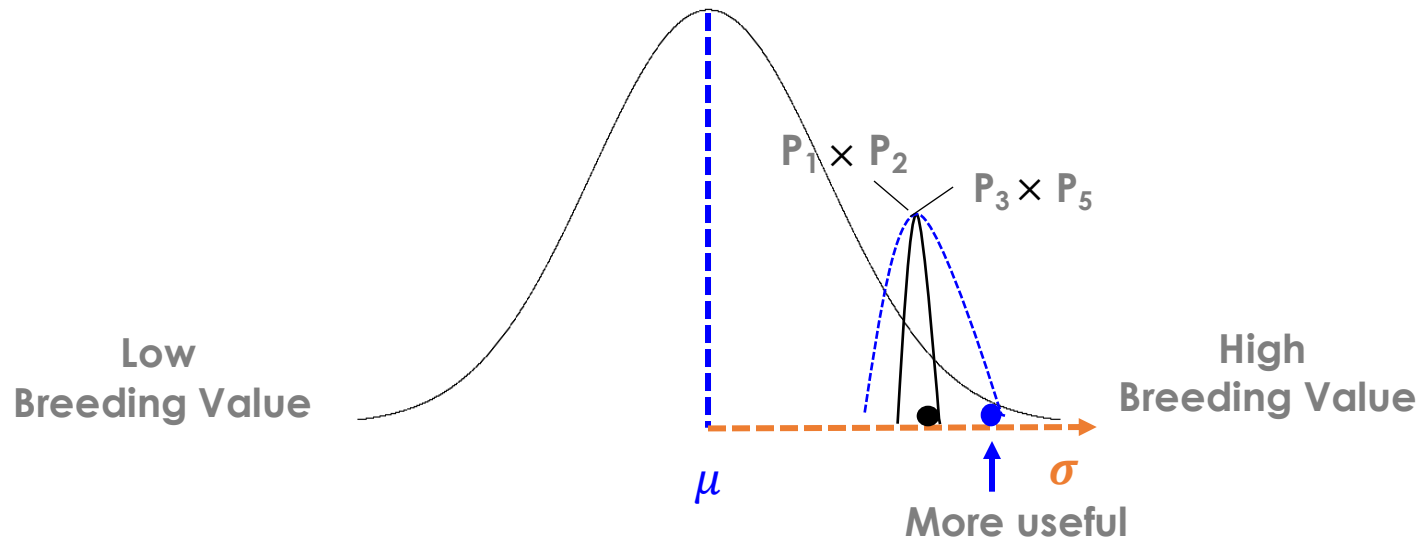
$$\mu_{gv} = \sum_{i=1}^n [a_i(p_i - q_i - y_i) + d_i[2p_iq_i + y_i(p_i - q_i)]]$$

4

Multi-trait for all above

SimpleMating – Cross performance

Prediction of cross performance: Usefulness



$$UC = \mu + i\sigma$$

μ = progeny mean

i = selection intensity

σ = progeny standard deviation

Schnell and Utz (1975)

Prediction of cross performance

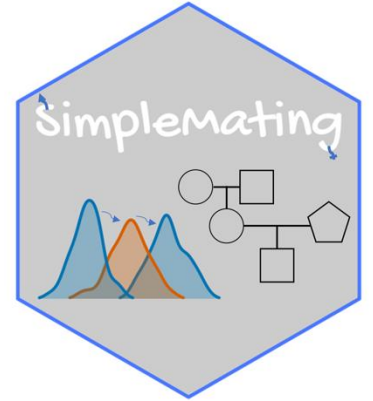
Output

```
### ----- 1. Performance prediction
```

```
ST_tgv2 <- getTGV(MatePlan = CrossPlan,  
                  Markers = Markers,  
                  addEff = addEff,  
                  domEff = domEff,  
                  K = relMat,  
                  ploidy = 2)
```

```
### ----- 1. Performance prediction
```

```
usefAdd <- getUsefA(MatePlan = CrossPlan,  
                    Markers = Markers,  
                    addEff = addEff,  
                    K = relMat,  
                    Map.In = Map, # Genetic map  
                    linkDes = NULL,  
                    propSel = 0.05,  
                    Type = "DH",  
                    Generation = 1)
```



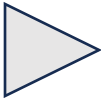
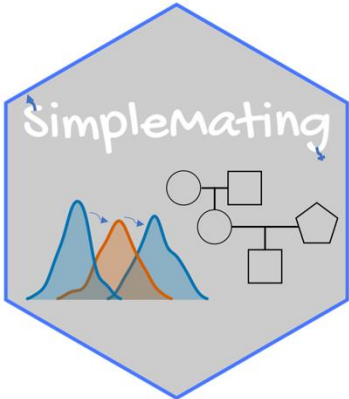
Output

```
### ----- 2. Optimization
MatePlan = selectCrosses(data = ST_tgv4,
                        n.cross = 20,
                        max.cross = 2,
                        min.cross = 1,
                        culling.pairwise.k = 0)
```

```
# stats
MatePlan[[1]]

# Plan
MatePlan[[2]]

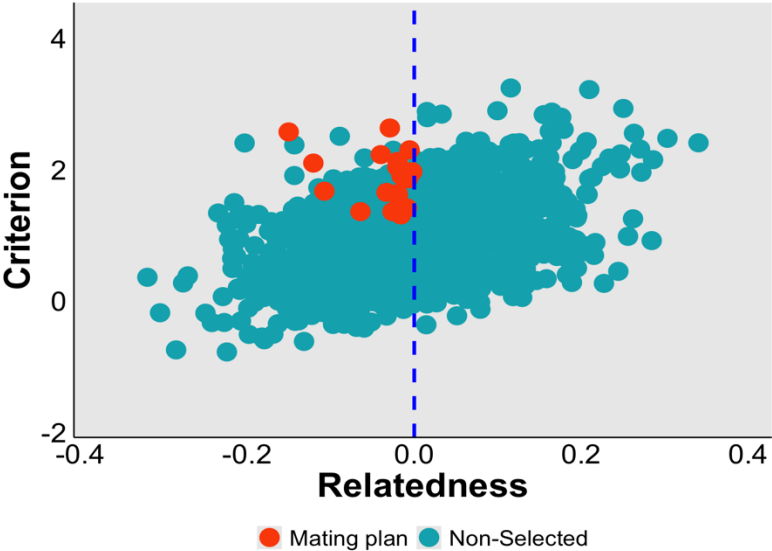
# Plot
MatePlan[[3]]
```



Criteria

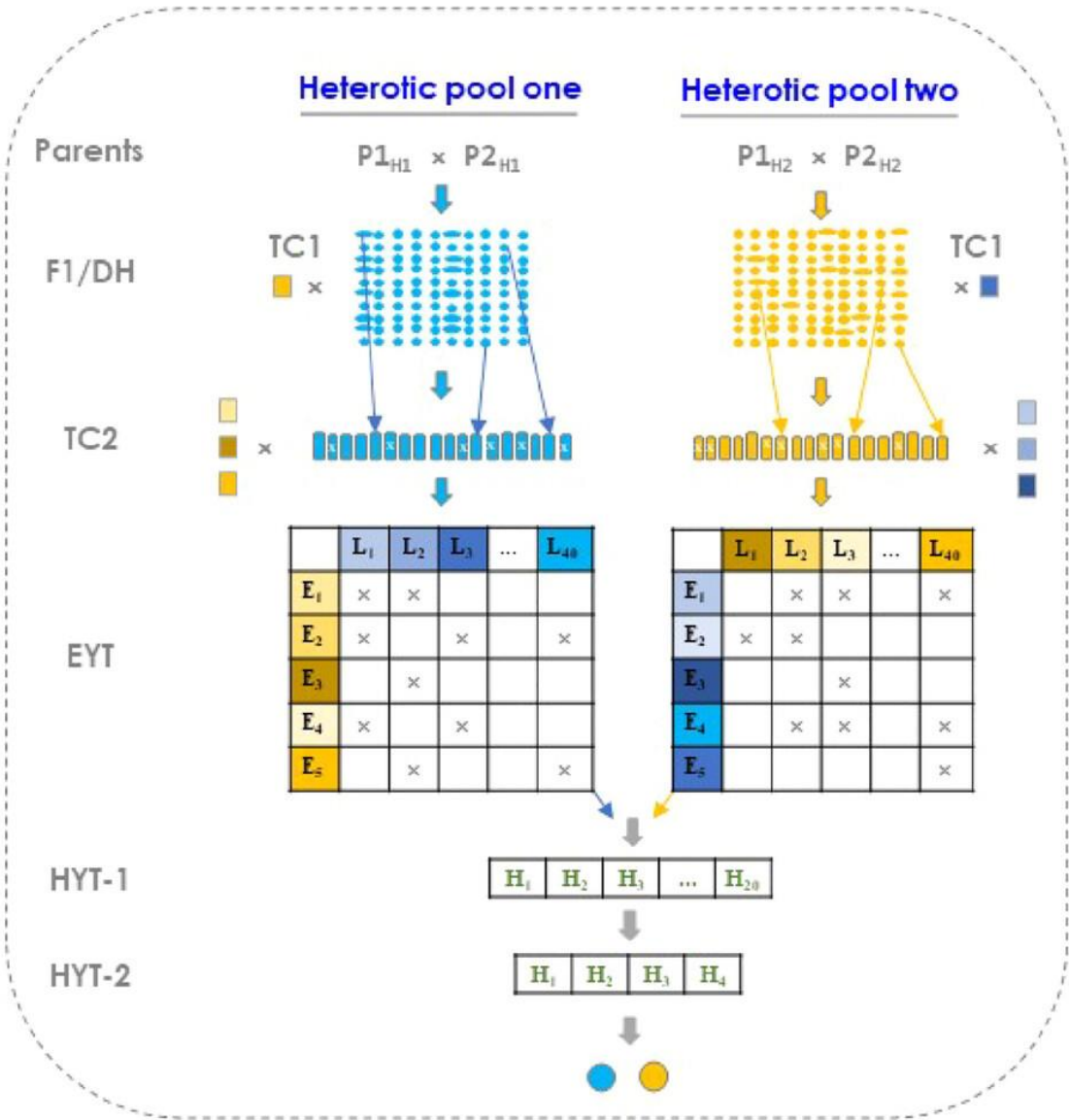
	P1	P2	Y	K
74	G58	G64	14.55678	-0.11572192
81	G61	G99	12.30788	-0.06813125
79	G61	G64	10.95003	-0.02893493
97	G8	G89	9.72175	-0.01851278
96	G8	G18	9.71596	-0.01758385
28	G23	G81	9.52128	-0.09772107
41	G33	G82	9.36602	-0.01468378
26	G23	G58	9.18741	-0.08279022
32	G27	G68	9.14856	-0.10668185
6	G11	G36	9.02959	-0.06179866

Mate plan Relationship

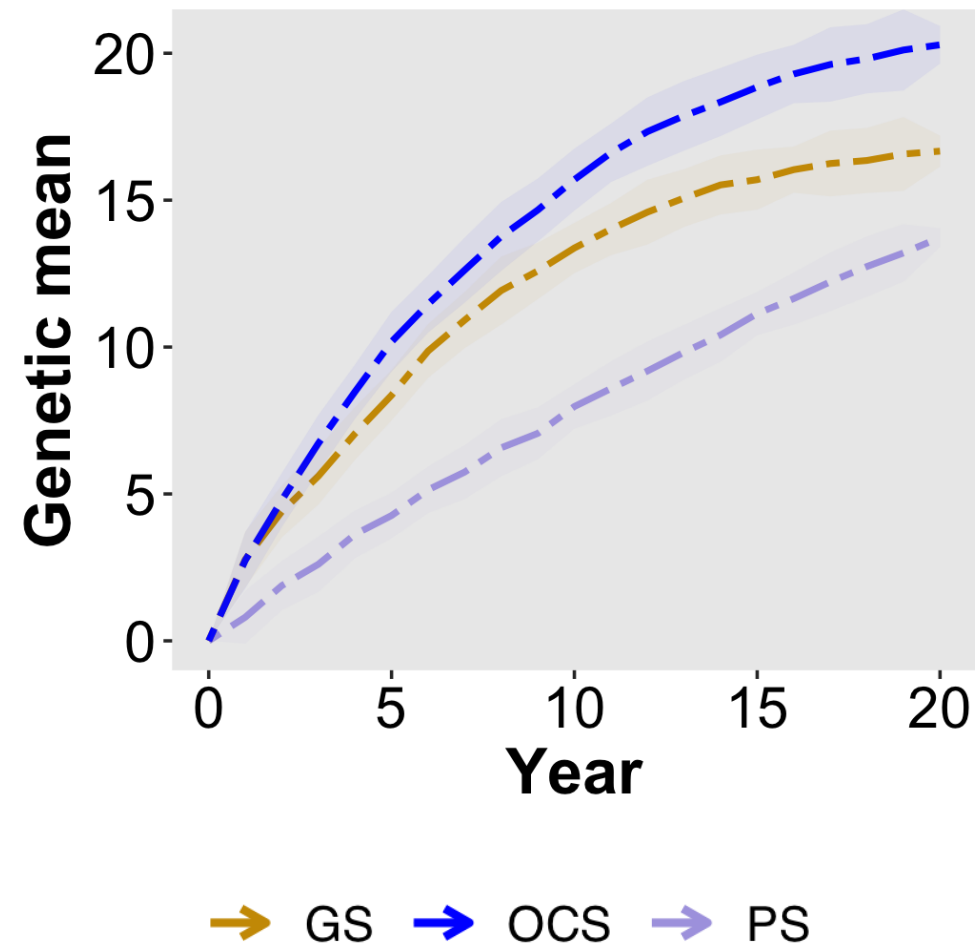


Example one: Hybrid breeding program

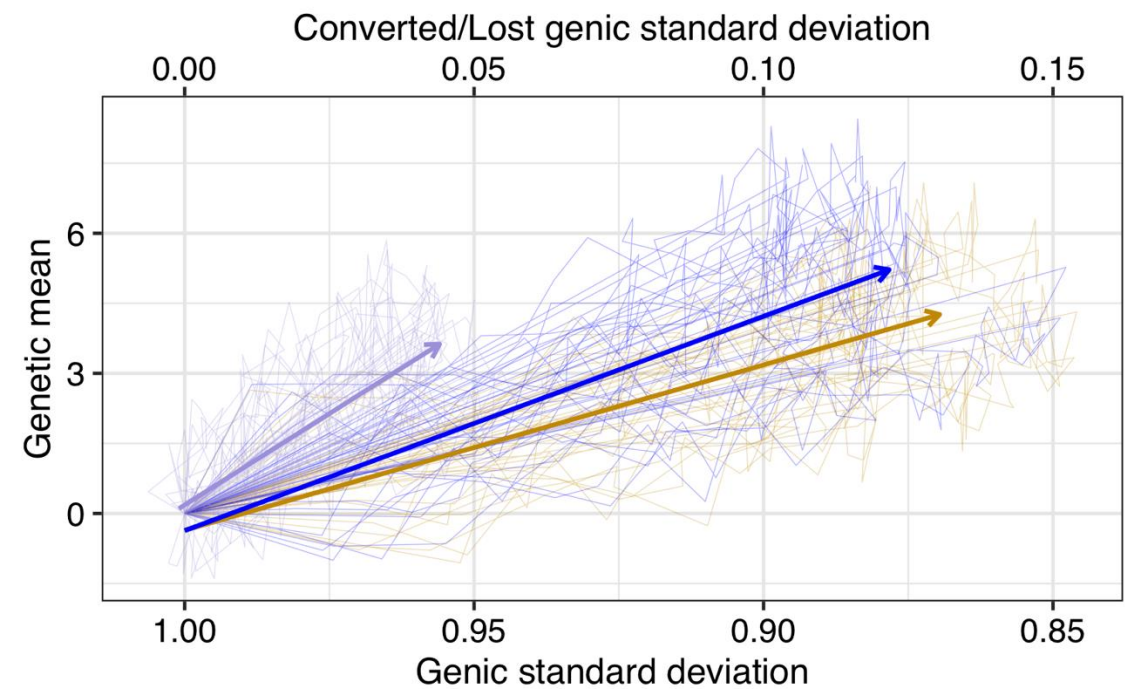
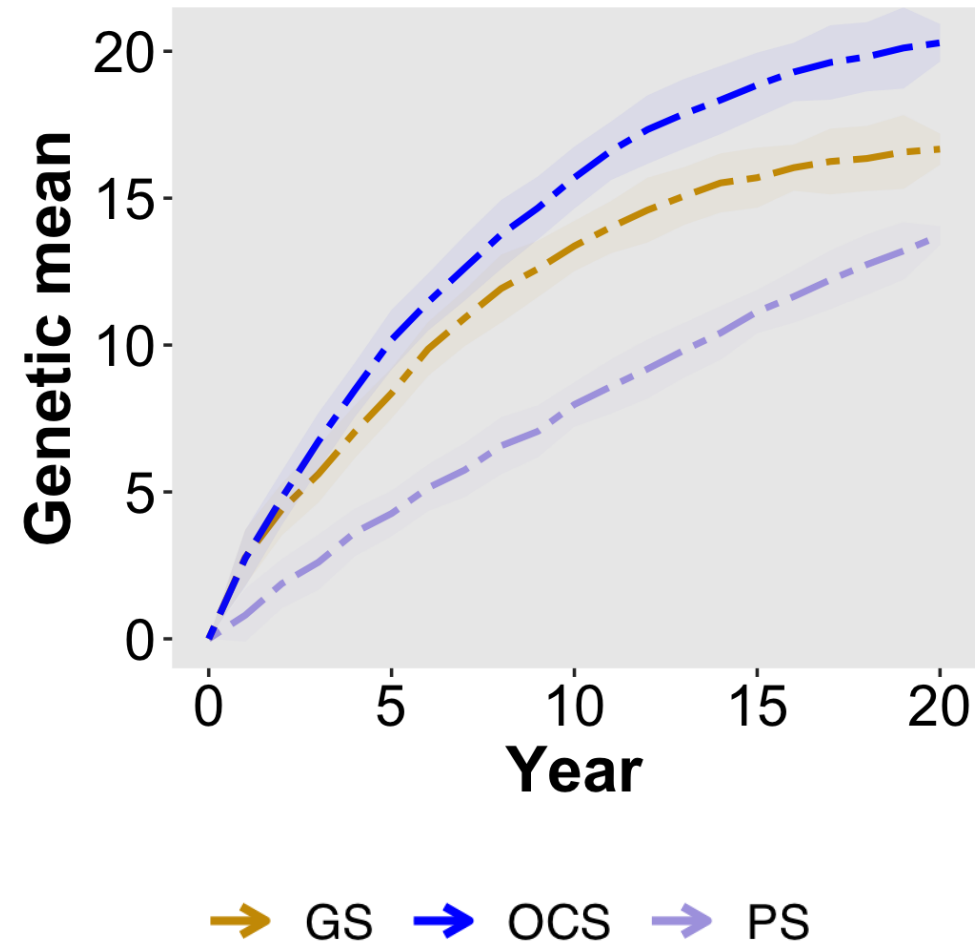
- Maize breeding program with two heterotic groups;
- Recurrent reciprocal selection scheme;
- PS: selection only using phenotypes;
- GS: Truncated selection via estimated breeding values;
- OCS: Optimal cross selection with usefulness as criterion (SimpleMating);
- 20 years of simulations (AlphasimR).



Example one: Hybrid breeding program

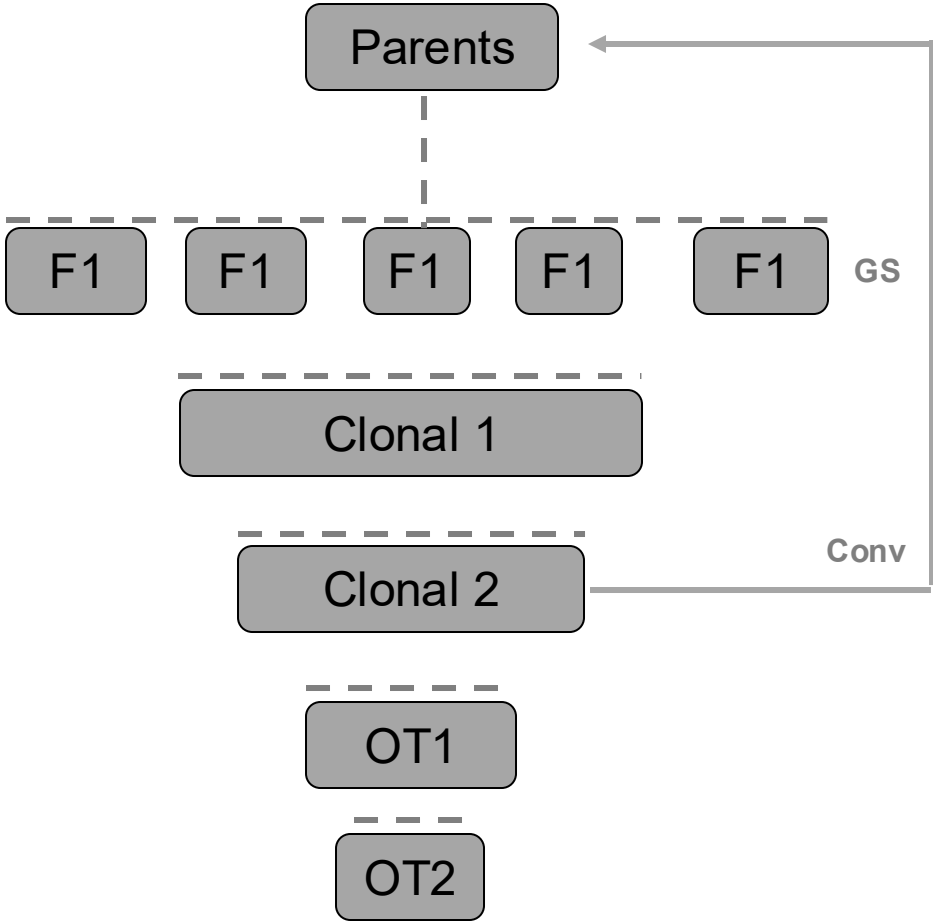


Example one: Hybrid breeding program

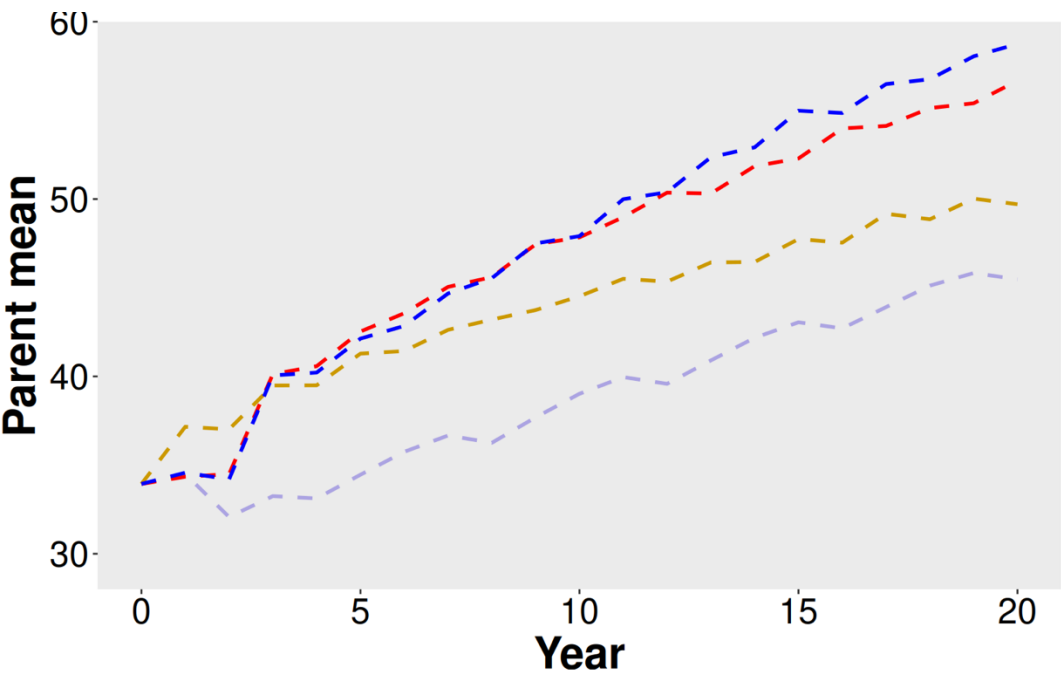


Example Two: Clonal breeding program

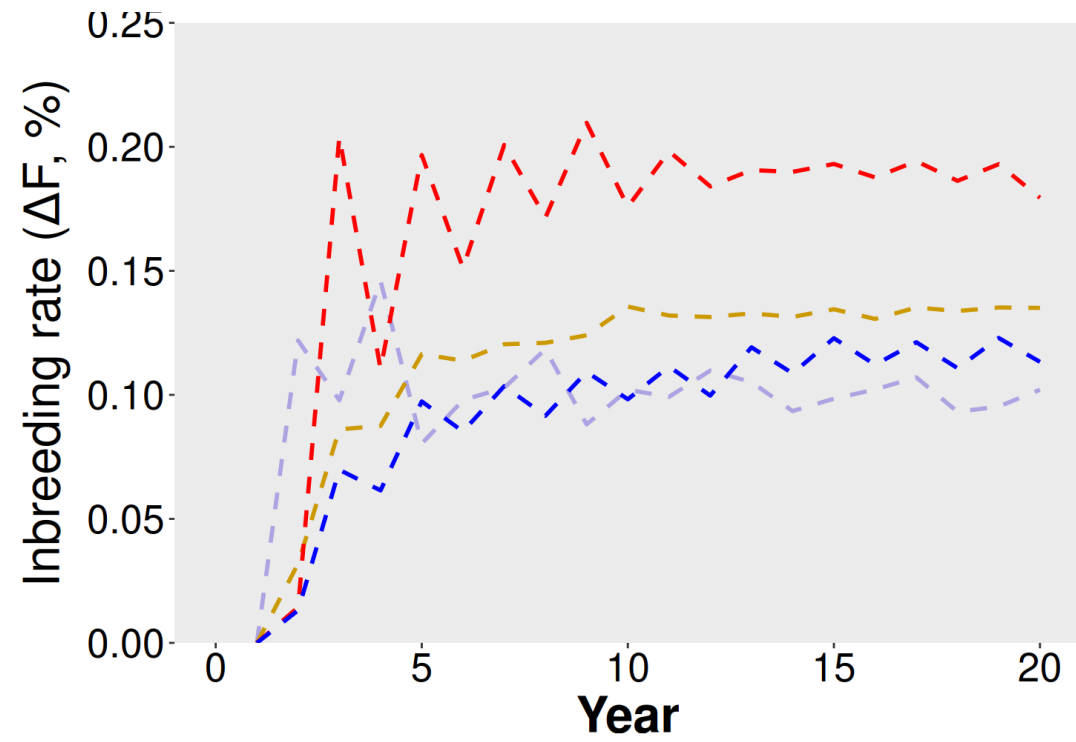
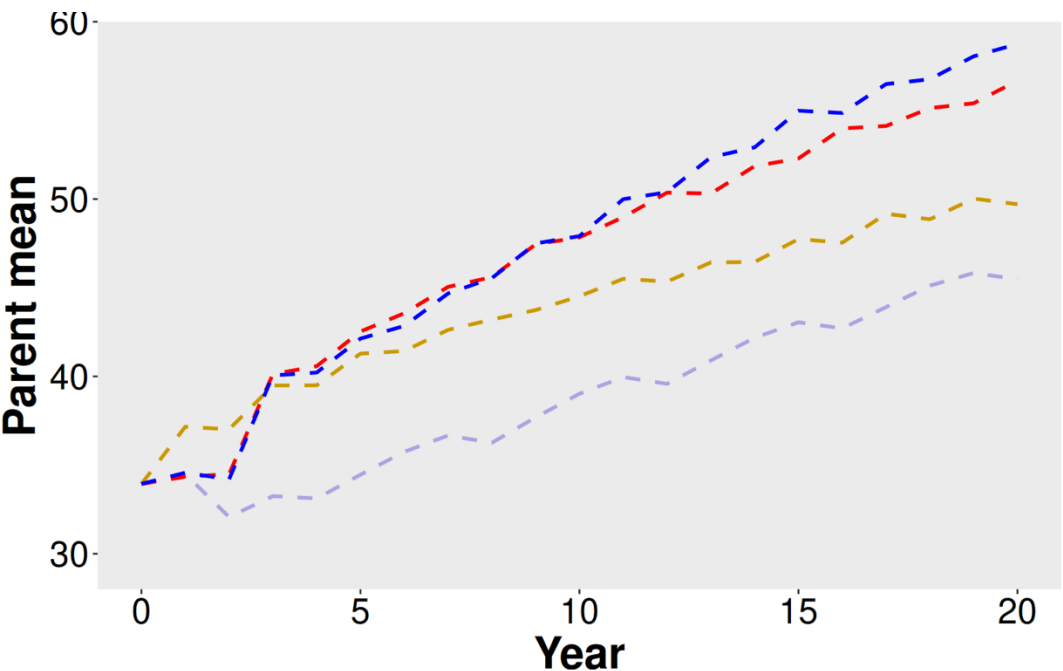
- Clonal breeding program (potato UF);
- Recurrent reciprocal selection scheme;
- PS: selection only using phenotypes;
- GS: Truncated selection via estimated breeding values;
- TGV: Truncated selection via total genetic value (A+D);
- OCS: Optimal cross selection with TGV as criterion (SimpleMating);
- 20 years of simulations (AlphasimR);
- Dominance deviation: 0.4.



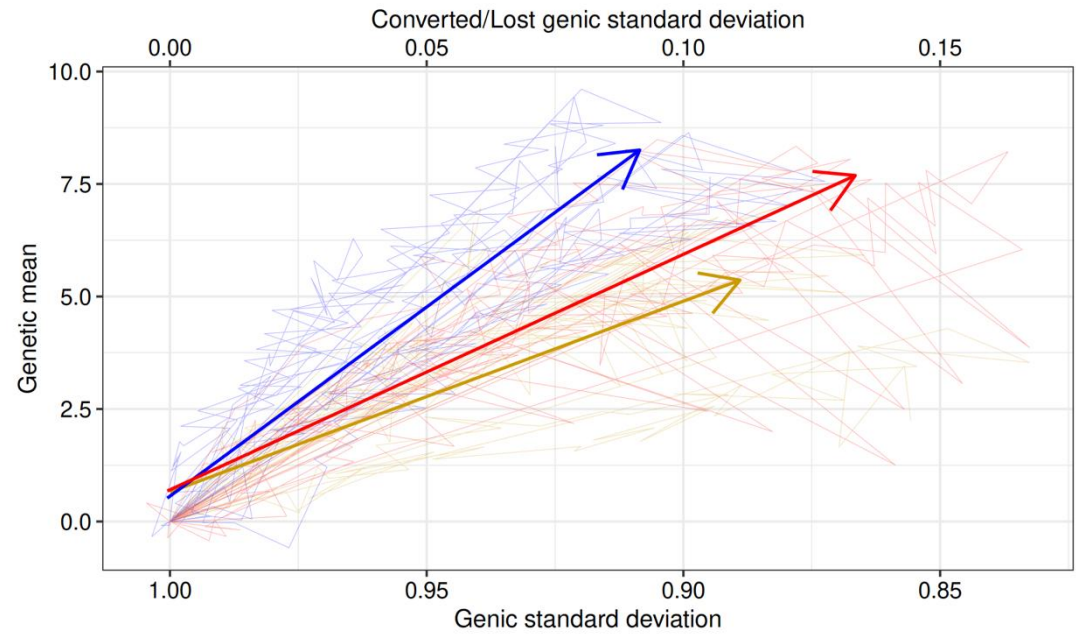
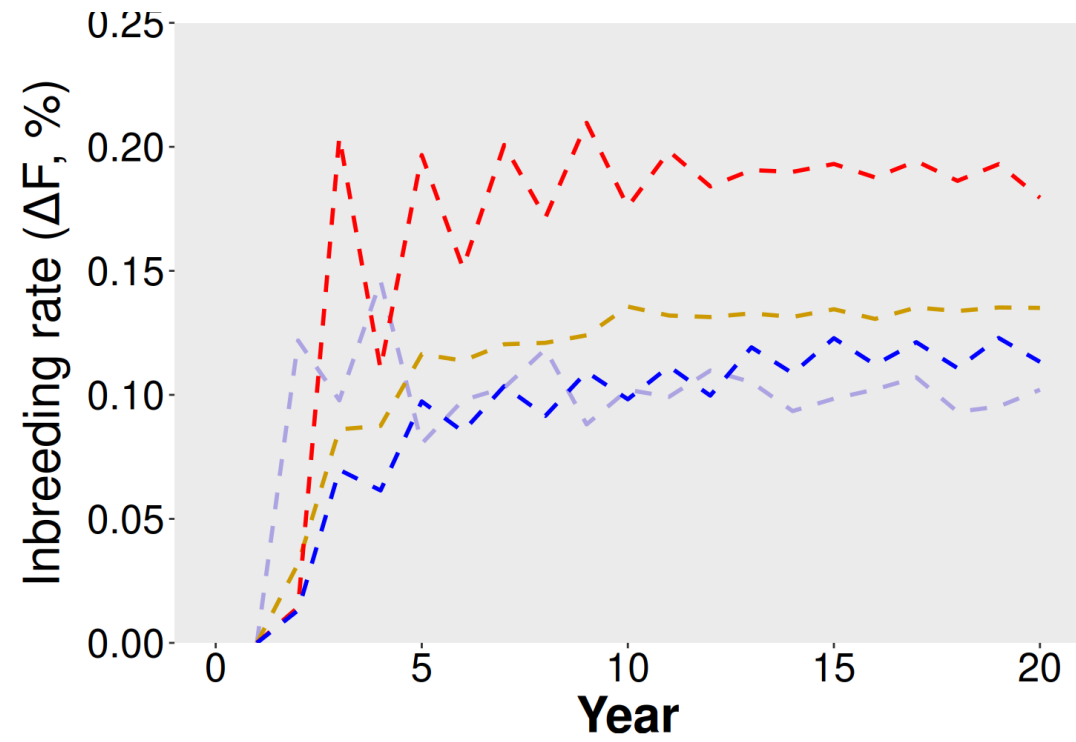
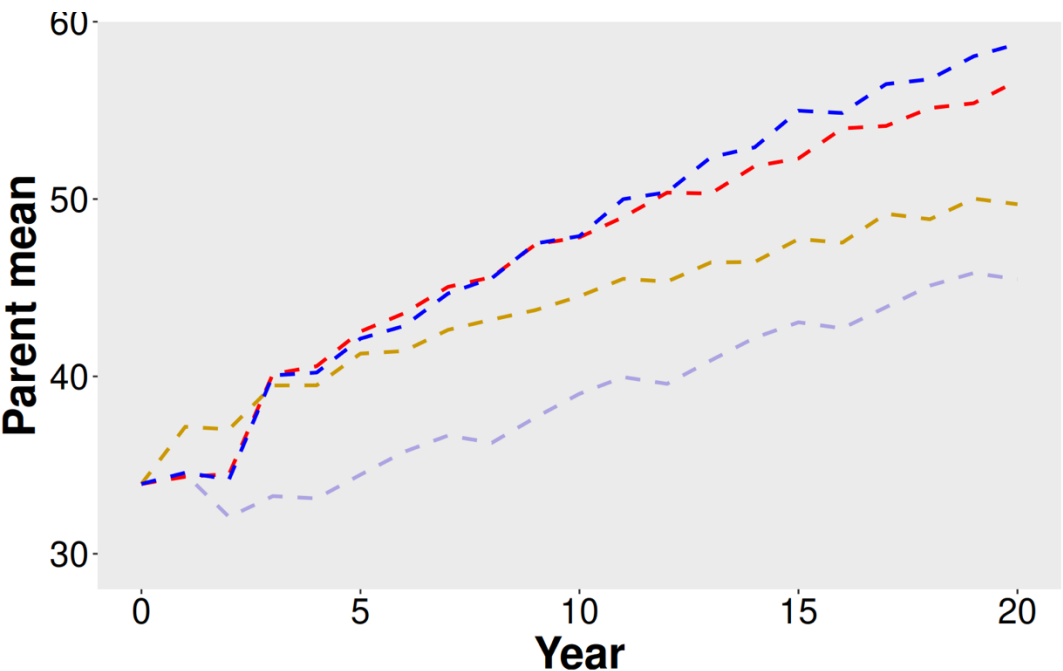
Example Two: Clonal breeding program



Example Two: Clonal breeding program

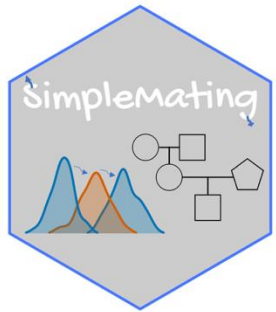


Example Two: Clonal breeding program





Use of **optimal cross selection** is an efficient implementation in various genomic selection-based breeding programs



SimpleMating provides an efficient framework for optimizing crosses and improving selection decisions in breeding programs

Acknowledgments

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<https://marcopxt.github.io/>

